Original Research Article

Disentangling the Associations Between Parental BMI and Offspring Body Composition Using the Four-Component Model

DELAN DEVAKUMAR, 1* CARLOS GRIJALVA-ETERNOD, 1 MARIO CORTINA-BORJA, 2 JANE WILLIAMS, 3 MARY FEWTRELL,³ AND JONATHAN WELLS³
¹UCL Institute for Global Health, London WC1N 1EH, United Kingdom 2 Clinical Epidemiology, Nutrition and Biostatistics Section, UCL Institute of Child Health, London WC1N 1EH, United Kingdom

 3 Childhoo \widetilde{d} Nutrition Research Centre, UCL Institute of Child Health, London WC1N 1EH, United Kingdom

ABSTRACT: Objectives: This study sets out to investigate the intergenerational associations between the body mass index (BMI) of parents and the body composition of their offspring.

Methods: The cross-sectional data were analyzed for 511 parent–offspring trios from London and south-east England. The offspring were aged 5–21 years. Parental BMI was obtained by recall and offspring fat mass and lean mass were obtained using the four-component model. Multivariable regression analysis, with multiple imputation for missing paternal values was used. Sensitivity analyses for levels of non-paternity were conducted.

Results: A positive association was seen between parental BMI and offspring BMI, fat mass index (FMI), and lean mass index (LMI). The mother's BMI was positively associated with the BMI, FMI, and LMI z-scores of both daughters and sons and of a similar magnitude for both sexes. The father's BMI showed similar associations to the mother's BMI, with his son's BMI, FMI, and LMI z-scores, but no association with his daughter. Sensitivity tests for non-paternity showed that maternal coefficients remained greater than paternal coefficients throughout but there was no statistical difference at greater levels of non-paternity.

Conclusions: We found variable associations between parental BMI and offspring body composition. Associations were generally stronger for maternal than paternal BMI, and paternal associations appeared to differ between sons and daughters. In this cohort, the mother's BMI was statistically significantly associated with her child's body composition but the father's BMI was only associated with the body composition of his sons. Am. J. Hum. Biol. 00:000–000, 2016. © 2016 The Authors American Journal of Human Biology Published by Wiley Periodicals, Inc.

The association between the body mass index (BMI) of parents and offspring has been of interest in relation to the worldwide rising trend of obesity. It is generally understood that BMI tracks from one generation to the next (Reilly et al., 2005) and that parental obesity strongly predicts offspring childhood obesity (Parsons et al., 1999).

However, several issues remain poorly understood. There is little information on what underlies such intergenerational correlations in BMI at the level of specific tissues, such as lean mass (LM) and fat mass (FM). While BMI is a useful indicator at a population level, it is a poor predictor of adiposity, as it incorporates both FM and LM but cannot differentiate their ratio (Wells, 2000; Wells et al., 2006). Thus, parent-offspring correlations might reflect consistency in adiposity, or in frame size and muscularity. The evidence is mixed as to whether a sex- or gender-specific association between parent and child exists, as discussed in more detail below, and a number of different mechanisms may be involved, with both environmental (Power et al., 2011) and genetic factors (Schousboe et al., 2003) transmitting parental effects to the child's phenotype. Indeed, each parent might influence its offspring through different behavioral and biologically mediated mechanisms, with boys and girls responding to different cues.

The possibility that parental BMI may show different associations with the phenotype of sons and daughters can be disentangled into two discrete components, and again a number of different mechanisms may be relevant. First, paternal and maternal BMI might show contrasting associations with child phenotype. The genetic basis of obesity is assumed to derive from the accumulation of risk alleles,

where each risk allele contributes a relatively small effect to BMI (Elks et al., 2012). In general, Mendelian inheritance is not sex specific; however, genetic imprinting is one pathway by which parents can exert different influences on their offspring. In this case, the allele from one parent is active, while the other is switched off. The placenta is thought to be particularly important for the control of imprinting, either as a temporary or permanent phenomenon (Frost and Moore, 2010). Programming effects may also occur. Stronger associations between maternal than paternal BMI and offspring BMI, reported in some but not all studies (Danielzik et al., 2002; Lawlor et al., 2007; Linabery et al., 2013; Whitaker et al., 2010), may indicate the influence of maternal physiology during pregnancy or lactation. However, imprinting of the sperm is also possible, and exposures such as paternal famine or smoking during adolescence have been associated with offspring phenotype (Northstone et al., 2014; Pembrey et al., 2006). Finally, the two parents might generate contrasting behavioral influences on offspring lifestyle. For instance,

Published online 00 Month 2016 in Wiley Online Library (wileyonlinelibrary.com).

This is an open access article under the terms of the [Creative Commons](http://creativecommons.org/licenses/by/3.0/) [Attribution](http://creativecommons.org/licenses/by/3.0/) License, which permits use, distribution and reproduction in any medium, provided the original work is properly cited.

Additional Supporting Information may be found in the online version of this article.

^{*}Corresponding author: Delan Devakumar, UCL Institute for Global Health, 30 Guilford St, London WC1N 1EH, United Kingdom. E-mail: d.devakumar@ucl.ac.uk

Received 27 April 2015; Revision received 15 August 2015; Accepted 8 December 2015

DOI: 10.1002/ajhb.22825

maternal attitudes, such as disinhibition, are known to influence child eating behaviors (Cutting et al., 1999).

Second, sons and daughters might respond in contrasting ways to parental phenotype. Even during early childhood, boys within the normal range of nutritional status tend to have greater fat-free mass than girls (Wells et al., 2012), while the tissue composition of excess weight gain also differs between the sexes (Wells et al., 2006). Such differences may relate to sex-different sensitivity to hormones such as insulin and leptin (Wells and Cole, 2014). In terms of developmental mechanisms, differential in utero DNA methylation of the fetus has been shown between the sexes (Cooper et al., 2012; Khulan et al., 2012). Evidence showing that paternal BMI is associated with the in utero growth of male offspring but not female offspring, supports this view (Chen et al., 2012; Pomeroy et al., 2015). It is also possible that mitochondrial inheritance (via the mother) (Giles et al., 1980) may play a role and elevated mother–child risks might seem to suggest this option, but the available evidence for obesity is not consistent to date (Barrett, 2001; Knoll et al., 2014; Yang et al., 2007).

Adding these two contrasting effects together, we hypothesize that mothers and fathers might generate different effects on sons compared to daughters. In this article, we investigate the intergenerational transmission of obesity by examining the association between parental BMI and their offspring's body composition. Body composition was assessed using the four-component (4C) model, considered the most accurate in vivo approach (Wells et al., 2012) because assumptions regarding the composition of the LM are minimal (Fuller et al., 1992; Wells et al., 2012). We tested whether associations between parental BMI and offspring FM or LM are associated with the sex of the parent or the offspring.

METHODS

We used a database of 533 children originally recruited to generate reference curves of body composition (Wells et al., 2012). Healthy children and young adults were recruited for the study from the general population using flyers and newspaper advertisements in London and the southeast of England from 2001. The age range was 4 to 23 years (young adults were included to cover the entire pediatric age range). There were no exclusion criteria for BMI. The sample was narrowed to 511 children aged 5–21 years, to match the age range for which body composition reference curves exist (we used the 20-year-old reference ranges for offspring aged 21 years). The methods are described elsewhere in detail (Wells et al., 2012). Briefly, all individuals attended Great Ormond Street Hospital in London for body-composition investigation. Weight and height were measured by using standard protocols, 60% of the measurements were by one operator and the rest by five other operators. FM and LM were calculated using the 4C model which utilizes body volume (BV) by airdisplacement plethysmography (Bodpod; Cosmed, Italy), bone mineral content from dual-energy X-ray absorptiometry (DXA; GE Medical Systems, UK) and total body water (TBW) by deuterium dilution. The technical error of measurement (TEM) for height was 0.38 cm and TEM% was 0.23% (Ulijaszek and Kerr, 1999). Bone mineral content (BMC) in Equations (1) and (2) were used to calculate FM and LM.

$$
FM = 2.747 \times BV - 0.710 \times TBW + 1.460 \times BMC - 2.050 \times Weight
$$
 (1)

$$
LM = Weight - FM \tag{2}
$$

Current parental height and weight were obtained by recall. The attending parent, usually the mother, was asked for her height and weight, and that of the other parent.

Ethical approval for the original study was granted by the Research Ethics Committee of University College London Institute of Child Health and Great Ormond Street Hospital.

Analysis

BMI was estimated as weight (in kg) divided by squared height (in meters). LM index (LMI) and FM index (FMI) were calculated based on the relationship described in Eqs. (3) and (4) .

$$
\frac{\text{Weight}}{\text{Height}^2} = \frac{\text{FM}}{\text{Height}^2} + \frac{\text{LM}}{\text{Height}^2}
$$
(3)

$$
BMI = FMI + LMI \tag{4}
$$

Exposure variables were parental BMI and outcome variables were the offspring anthropometry z-scores. Offspring BMI, LMI, and FMI were converted to age and sexspecific z-scores based on the British 1990 growth references (BMI) and the new UK body composition reference child (LMI and FMI) using the lmsGrowth Excel add-in (version 2.74) (Cole et al., 1995; Wells et al., 2012).

Associations were initially examined using correlations and then univariable and multivariable linear regression. First, the association of maternal and paternal BMI was compared with each of the offspring outcome variables for all offspring together, and then stratified by sex. The multivariable model included the other parent's BMI to account for assortative mating, where men and women do not mate at random as has been shown for fat and lean mass (Speakman et al., 2007). We calculated the variance inflation factor for these models using the function \forall if from R library car (Fox and Weisberg, 2011).

The data were initially limited to those for which there were complete data for both parents and offspring. Missing data points comprised approximately 14% of the paternal data, therefore we decided to impute values for the parent's BMI, using multiple imputation with chained equations (MICE) to improve our inferences on the associations between the exposure and the outcome variables (van Buuren, 2012). The offspring data were complete. We checked the assumption of data missing completely at random for every model, performing t -tests on the response variable in each regression model against missing/not missing for each explanatory variable in the model. We considered 10 multiple imputations for each model.

Additional sensitivity analyses were then done to take into account potential non-paternity using Steer's correction to Clemons' formula: (Clemons, 2000; Steer, 2009) Let σ_{mm} , and σ_{ff} be the variances of the maternal and reported paternal BMI, and σ_{fm} and σ_{bm} be the covariances between

| Variable | | Boys | Girls | | |
|---|---------------|------------------------------------|---------------|------------------------------------|--|
| Number | | 248 | 263 | | |
| Ethnicity | | $n(\%)$ | n(%) | | |
| European | | 221 (89.1%) | $242(92.0\%)$ | | |
| Asian | | $8(3.2\%)$ | $2(0.8\%)$ | | |
| Black | | $10(4.0\%)$ | $12(4.6\%)$ | | |
| Chinese | | $3(1.2\%)$ | $4(1.5\%)$ | | |
| Other | $6(2.4\%)$ | | $3(1.1\%)$ | | |
| | Mean(SD) | Range | Mean(SD) | Range | |
| Age (years) | 12.3(4.0) | 5.1 to 21.8 | 12.8(4.3) | 5.0 to 21.9 | |
| Weight (kg) | 45.2(19.3) | 16.1 to 106.4 | 46.1(16.4) | 17.0 to 90.8 | |
| Weight z -score | 0.29(1.08) | -2.42 to 3.44 | 0.42(1.10) | -2.75 to 3.46 | |
| Height (cm) | 150.7(20.5) | 104.7 to 190.4 | 149.9(17.8) | 107.0 to 181.7 | |
| Height z -score | 0.20(0.96) | -2.09 to 3.28 | 0.35(1.03) | -2.77 to 3.42 | |
| BMI (kg/m ²) | 18.9(3.9) | 13.0 to 37.9 | 19.8(3.8) | 12.8 to 34.5 | |
| $BMI z-score$ | 0.22(1.16) | -2.99 to 3.49 | 0.31(1.14) | -3.33 to 3.09 | |
| Fat mass (kg) | 9.1(6.8) | $1.0 \text{ to } 45.5$ | 13.1(7.6) | $2.3 \text{ to } 40.0$ | |
| FMI z-score | 0.01(0.99) | -3.28 to 2.34 | 0.00(1.00) | -2.41 to 2.41 | |
| Lean mass (kg) | 36.1(15.0) | 13.7 to 76.4 | 33.0(10.2) | 13.2 to 61.9 | |
| $LMI z-score$ | $-0.01(1.08)$ | -4.3 to 5.13 | 0.01(1.10) | -3.23 to 2.77 | |
| Mother's BMI (kg/m^2) | 25.5(4.7) | 17.0, 21.7, 23.3, 25.2, 29.5, 41.9 | 25.3(5.2) | 16.0, 21.6, 23.2, 25.0, 28.5, 43.9 | |
| $(Range = Minimum, 1st to)$ 4th quintiles, maximum) | | | | | |
| Father's BMI (kg/m^2) $(Range = Minimum, 1st to)$ 4th quintiles, maximum) | 26.5(3.7) | 16.9, 23.5, 25.1, 27.3, 29.7, 37.3 | 26.3(4.1) | 17.7, 22.8, 24.7, 26.7, 29.3, 37.4 | |

TABLE 1. Participant characteristics

the maternal and reported paternal BMI, and the maternal and biological paternal BMI. Let p be the probability that the reported father is not the biological father, then α is an indicator function with value $(1-p)$ if $\sigma_{bm}=0$, and 1 if $\sigma_{bm} = \sigma_{\text{fm}}$. The attenuation matrix can be written as:

$$
\Lambda = \begin{pmatrix} \sigma_{ff} & a \sigma_{ff} \\ a \sigma_{ff} & \sigma_{ff} \end{pmatrix}^{-1} \begin{pmatrix} (1-p)\sigma_{ff} & (1-p)\sigma_{fm} \\ \sigma_{ff} & \sigma_{mm} \end{pmatrix}
$$
 (5)

If the maternal and paternal regression coefficients are β_1 and β_2 then the coefficients adjusted for potential nonpaternity are:

$$
\Lambda \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix} = \begin{pmatrix} \beta_1^* \\ \beta_2^* \end{pmatrix} =
$$
\n
$$
\begin{pmatrix} \frac{(-1+a+p)\beta_2\sigma_{fm}\sigma_{mm} + \beta_1 \left(a(-1+p)\sigma_{fm}^2 + \sigma_{ff}\sigma_{mm} \right)}{(-1+p)\left(\sigma_{fm}^2 - \sigma_{ff}\sigma_{mm} \right)} \\ - \frac{(1+a(-1+p))\beta_1\sigma_{ff}\sigma_{fm} + \beta_2 \left(a\sigma_{fm}^2 + (-1+p)\sigma_{ff}\sigma_{mm} \right)}{(-1+p)\left(\sigma_{fm}^2 - \sigma_{ff}\sigma_{mm} \right)} \end{pmatrix} (6)
$$

We used $a = 1$ as, in the absence of any further data, we considered that assuming equality of correlation (reflected in the covariance values) for the association patterns present (i) between maternal BMI and biological father's BMI and (ii) between maternal BMI and reported father's BMI was more informative than assuming lack of correlation between maternal BMI and biological father's BMI. The variances of the parental heights were calculated applying Rubin's rule to the imputed datasets, (Rubin, 1987) which gives a weighted average of the estimates obtained by fitting the regression models to each of the 10 imputed datasets.

We tested the hypothesis of no difference between maternal and paternal regression coefficients on offspring BMI, FMI, and LMI z-scores for different assumed percentages of non-paternity using a bootstrap procedure which took into account the MICE models. The one-sided p-values correspond to the alternative hypothesis of maternal effects being larger than paternal effects, and were obtained analyzing 10,000 realizations of the empirical resampling distribution of these differences.

Statistical analyses were carried out using Stata software version 12.1 (Stata Corporation, Texas 2011) and in the R language and environment for statistical computing version 3.0.1 (Team, 2011). Regression models accounting for missing values were fitted in R using the package mice (van Buuren and Groothuis-Oudshoorn, 2011) to perform MICE.

RESULTS

A description of the sample is given in Table 1. Ninetyone percent of the offspring reported to be ethnically European, the remainder were of varied ethnic origin, including mixed ethnic origin. The mean height and weight z-scores were greater than zero, with a range of >5 z-scores. Girls were a little taller than boys but also presented more variability in their height $(t$ test for difference between girls and boys: weight-for-age $p = 0.18$, height-for-age $p = 0.09$). The Pearson correlation coefficient between maternal and paternal BMI was 0.143 $(p = 0.003)$. The variance inflation factor was 1.021, which did not suggest concerns regarding collinearity.

The data are shown in scatterplots for each parent separately in Figures 1a,b. When considering sons and daughters together, there was a positive association between both parent's BMI and offspring BMI, FMI and LMI (Table 2). This association was present in both the univariable and adjusted model (Table 2 and Supporting Information Table 1a) and was present with and without imputation (Supporting Information Table 1b). The regression coefficients for

Fig. 1. Scatterplots to show the maternal (1a) and paternal BMI (1b) and offspring body composition.

| | | Predicted change in offspring body composition, β (95% CI), associated with parental BMI ^a | | | | | |
|-----------------------------------|--|--|---|--|--|--|--|
| | | Mother's BMI | | Father's BMI | | | |
| Offspring body composition | | Unadjusted | Adjusted ^b | Unadjusted | Adjusted ^c | | |
| All offspring Female offspring | $BMI z-score$ FMI z-score LMI z -score $BMI z-score$ FMI z-score | $0.078(0.059 \text{ to } 0.097)$ $0.061(0.044 \text{ to } 0.079)$ $0.053(0.034 \text{ to } 0.071)$ $0.075(0.049 \text{ to } 0.100)$ $0.057(0.033 \text{ to } 0.080)$ | $0.073(0.053 \text{ to } 0.093)$ 0.056(0.039 to 0.074) 0.049(0.030 to 0.067) $0.071(0.044 \text{ to } 0.097)$ 0.055(0.031 to 0.078) | $0.052(0.024 \text{ to } 0.080)$ $0.031(0.006 \text{ to } 0.056)$ $0.036(0.012 \text{ to } 0.061)$ $0.032(-0.004 \text{ to } 0.069)$ $0.019(-0.020 \text{ to } 0.058)$ | $0.038(0.011 \text{ to } 0.065)$ 0.029(0.007 to 0.051) $0.029(0.003 \text{ to } 0.055)$ $0.029(-0.005 \text{ to } 0.063)$ $0.017 (-0.017 \text{ to } 0.051)$ | | |
| Male offspring | LMI z -score $BMI z-score$ FMI z-score LMI z -score | $0.060(0.037 \text{ to } 0.083)$ 0.081(0.051 to 0.110) $0.068(0.043 \text{ to } 0.094)$ $0.043(0.013 \text{ to } 0.072)$ | $0.057(0.033 \text{ to } 0.081)$ $0.074(0.045 \text{ to } 0.104)$ $0.064(0.038 \text{ to } 0.090)$ $0.040(0.010 \text{ to } 0.070)$ | $0.037(0.004 \text{ to } 0.070)$ $0.072(0.031 \text{ to } 0.112)$ $0.054(0.019 \text{ to } 0.090)$ 0.040(0.001 to 0.078) | $0.028(-0.004 \text{ to } 0.061)$ $0.057(0.018 \text{ to } 0.096)$ $0.041(0.006 \text{ to } 0.077)$ $0.029(-0.011 \text{ to } 0.070)$ | | |

TABLE 2. Sex-specific associations of offspring body composition (BMI, FMI, LMI) with parental BMI

a Results with multiple imputation, per one-unit increase. b Adjusted for father's BMI.

c Adjusted for mother's BMI.

Fig. 2. Associations of childhood z-scores for BMI, FMI, and LMI with Parental BMI: regression coefficients with 95% confidence intervals (with imputation).

BMI, LMI, and FMI for maternal BMI tended to be approximately one and a half to two times that of the paternal BMI. The associations of maternal and paternal BMI were similar for both FMI and LMI of the offspring.

When considering boys and girls separately, the association of the mother's BMI with the BMI of both boys and girls was statistically significant in all cases and of a similar magnitude for both. While not showing definitive evidence for this, there was a suggestion of a greater association with FMI than LMI in boys. In girls there was a marginally greater association with LMI. For the father's BMI there was only a significant association with the BMI, FMI and LMI z-scores of boys. The magnitude of the association for boys was similar to that of the mother's association. Again there was a suggestion of a greater association with FMI in boys and LMI in girls and this difference was maintained with imputed results. Of interest, significant associations were not seen between the father's BMI and any of the daughter's tested indices. The associations for each parent and child are shown in Figure 2.

Models fitted with multiple imputation showed a slight strengthening of the association of the mother's BMI and weakening of the father's BMI for all offspring outcome variables. It seemed a fair assumption that the missing data occurred completely at random. The p-values testing for a difference between the mean values of missing and not missing data showed an inability to reject the null hypothesis on most occasions, as shown in Supporting Information Table 2. With a p-value of 0.03, it is possible that the association between father's BMI and daughter's FMI did not achieve this criteria. We are aware, however, that it is not possible to directly test this assumption and that this seemingly significant result may have occurred by chance. We have therefore continued to use this imputed data, but treat this particular result with caution. The main conclusion from this approach was the same as in the non-imputed one.

The results for different levels of non-paternity are shown in Supporting Information Table 3. In each case, the maternal regression coefficient falls and the paternal rises with increasing levels of non-paternity. Maternal coefficients remained greater than paternal throughout, but the p-values indicated no statistical difference at greater levels of non-paternity.

DISCUSSION

Our analyses produced two important findings. First, overall we found a stronger association of maternal than paternal BMI with offspring body composition, though the strength of this effect varied across outcomes and was statistically significant in daughters only. Second, we found that the association between paternal phenotype and that of the offspring differed between sons and daughters, being significant in the former but not the latter. Therefore, both hypotheses are supported, in that the two parents generate different associations, and sons and daughters are also associated with their parents in different ways. Previous findings on these issues have been inconsistent, despite several studies on a very large sample size, hence we briefly review these findings.

BMI

A review of the literature by Patro et al. (2013) looked at the effect of prenatal maternal versus paternal BMI on offspring BMI or adiposity over the age of 5 years. They tended to find a greater maternal association with offspring BMI but showed inconsistent findings and daughters showed equal or different associations (Patro et al., 2013). Our findings are in keeping with studies by Lawlor et al. (2007), Whitaker et al. (2010), Danielzik et al. (2002), Linabery et al. (2013), and Svensson et al. (2011;

American Journal of Human Biology

TABLE 3. Continued TABLE 3. Continued

DISENTANGLING PARENTAL BMI AND OFFSPRING BODY COMPOSITION ASSOCIATIONS 7

at seven years of age) who showed a greater association with the mother's BMI than that of the father's. As described previously, this may be explained by biological or behavioral mechanisms, as mothers tend to be more involved in all aspects of child rearing.

We also showed the association of the father's BMI with sons was a little lower, but similar to that of the mother's, whereas there was no association between the father's BMI and that of the daughters. Regarding the father/ mother to daughter/son relationship, our results mirror Power et al. (2011) who found a lower correlation coefficient for father-to-daughter than mother-to-daughter or son or father-to-son. The results however differ from Burke et al. (2001) who suggested a greater association of father-to-daughter. Jääskeläinen et al. (2011) similarly showed a greater father-daughter association for parental BMI measured prior to the child's conception, but not when estimated at the same time as offspring measurement. The other studies in Table 3 tend to show no sexspecific difference or stronger associations between one parent and offspring (except Perez-Pastor et al. (2009) who show a stronger association between mother-todaughter and father-to-son).

Overall, no consistent pattern has emerged from research on this issue. Given the multiple different routes through which each parent can affect the phenotype of their offspring (Wells, 2014), it is very plausible that the association between parental phenotype and offspring BMI is genuinely heterogeneous across populations, as has been shown previously for the association between infant weight gain and later body composition (Wells et al., 2007).

Fat and lean mass

Our research goes further than most previous work in that it investigates whether an association exists between parental BMI and FM and LM in the offspring, using the 4C model. We found that the mother's BMI shows similar associations with that of her sons' and daughters' body composition. For fathers, no statistically significant associations were found with girls, in contrast to boys, where an association was seen with FM. Interestingly, among daughters there were similar associations for FM and LM, whereas in boys, with both parents, there was a greater association with FM than LM.

We found two studies that investigated associations between parental BMI and offspring FM and LM. Lawlor et al. (2008) assessed children at ages 9–11 years using DXA scans. They also found a strong association between parental BMI and offspring adiposity that was greater for mothers than fathers, and a weaker association between parental BMI and child LM that was of similar magnitude for each parent (Lawlor et al., 2008). In a cohort study from India, Veena et al. (2012) generally showed no difference between maternal adiposity during pregnancy or paternal adiposity when children were 5 years old and the adiposity of daughters or sons measured by bioelectrical impedance at age 9 years.

From an evolutionary perspective, differential effects of the two parents on the body composition of their offspring are expected to derive from fathers and mothers using different strategies to maximize their reproductive fitness. While the mother invests nutritional resources directly during early life, the father only invests information that may manipulate maternal physiology in his interests (Haig, 1993). Sex-different responses of the offspring are further predicted, because lean mass and fat mass bring different reproductive pay-offs for sons versus daughters (Wells et al., 2011).

Strengths

The main strength of our study was the use of the goldstandard 4C model to produce estimates of FM and LM in children. This allows us to predict more accurately the associations of parental BMI with different tissues. We have also accounted for assortative mating using a multiple regression model, and compensated for missing data by using multiple imputation. We have then tested our assumptions by giving estimates of the effects at different levels of non-paternity.

Limitations

The main limitation was the estimation of parental BMI by recall. Parental recall is likely to have resulted in error in the parental height and weight estimates, but a potential bias may work in either direction and would not explain the difference in results for sons and daughters, in that we would not anticipate recall error to be associated with offspring sex. Additionally, our findings are similar to some other studies that have measured parental anthropometry (Table 3). As described previously, several mechanisms exist by which environmental or genetic processes could allow parental phenotype to influence different body tissue types in sons and daughters. Further research is needed to elucidate in more detail the heterogeneity of findings on this research issue. We were unable to consider whether the association between the parent's BMI and offspring body composition varied within an individual over time. It is plausible that this association is different in early life than later in childhood and this is an important area of future research.

SUMMARY

Our work indicates a stronger association of maternal than paternal BMI with offspring body composition, and the suggestion that paternal associations may differ between sons and daughters. Given that parental effects on offspring BMI are established to be a strong predictor of childhood obesity (Parsons et al., 1999), our findings indicate that multiple and complex mechanisms contribute. Further work on this issue is required in order to improve understanding of how obesity prevention strategies may be developed to minimize the intergenerational transmission of childhood obesity.

ACKNOWLEDGMENTS

There was no specific funding for this research and no funder influenced this research. DD had a personal fellowship funded by the Wellcome Trust (092121/Z/10/Z). The MRC Centre of Epidemiology for Child Health is supported by funds from the UK Medical Research Council (grant G0400546). Research at the UCL Institute of Child Health & Great Ormond Street Hospital for Children receives a proportion of funding from the Department of Health's National Institute for Health Research Biomedical Research Centres funding scheme. The research study was designed by JCKW. Data were collected and body

composition variables modelled by JEW. The analysis was conducted by CGE, DD and MCB. The paper was drafted by DD with comments and amendments by all other authors.

REFERENCES

- Barrett TG. 2001. Mitochondrial diabetes, DIDMOAD and other inherited diabetes syndromes. Best Pract Res Clin Endocrinol Metab 15: 325–343.
- Burke V, Beilin LJ, Dunbar D. 2001. Family lifestyle and parental body mass index as predictors of body mass index in Australian children: a longitudinal study. Int J Obes Relat Metab Disord 25:147–157.
- Chen YP, Xiao XM, Li J, Reichetzeder C, Wang ZN, Hocher B. 2012. Paternal body mass index (BMI) is associated with offspring intrauterine growth in a gender dependent manner. PloS One 7:e36329.
- Clemons T. 2000. A look at the inheritance of height using regression toward the mean. Hum Biol 72:447–454.
- Cole TJ, Freeman JV, Preece MA. 1995. Body mass index reference curves for the UK, 1990. Arch Dis Child 73:25–29.
- Cooper WN, Khulan B, Owens S, Elks CE, Seidel V, Prentice AM, Belteki G, Ong KK, Affara NA, Constancia M, Dunger DB. 2012. DNA methylation profiling at imprinted loci after periconceptional micronutrient supplementation in humans: results of a pilot randomized controlled trial. FASEB J 26:1782–1790.
- Cutting TM, Fisher JO, Grimm-Thomas K, Birch LL. 1999. Like mother, like daughter: familial patterns of overweight are mediated by mothers' dietary disinhibition. Am J Clin Nutr 69:608–613.
- Danielzik S, Langnase K, Mast M, Spethmann C, Muller MJ. 2002. Impact of parental BMI on the manifestation of overweight 5-7 year old children. Eur J Nutr 41:132–138.
- Elks CE, den Hoed M, Zhao JH, Sharp SJ, Wareham NJ, Loos RJ, Ong KK. 2012. Variability in the heritability of body mass index: a systematic review and meta-regression. Front Endocrinol 3:29.
- Fleten C, Nystad W, Stigum H, Skjaerven R, Lawlor DA, Davey Smith G, Naess O. 2012. Parent-offspring body mass index associations in the Norwegian mother and child cohort study: a family-based approach to studying the role of the intrauterine environment in childhood adiposity. Am J Epidemiol 176(2):83–92.
- Fox J, Weisberg S. 2011. An R companion to applied regression, 2nd ed. Thousand Oaks, CA: Sage. Frost JM, Moore GE. 2010. The importance of imprinting in the human
- placenta. PLoS Genet 6:e1001015.
- Fuller NJ, Jebb SA, Laskey MA, Coward WA, Elia M. 1992. Four-component model for the assessment of body composition in humans: comparison with alternative methods, and evaluation of the density and hydration of fat-free mass. Clin Sci (Lond) 82:687–693.
- Giles RE, Blanc H, Cann HM, Wallace DC. 1980. Maternal inheritance of human mitochondrial DNA. Proc Natl Acad Sci USA 77:6715–6719.
- Haig D. 1993. Genetic conflicts in human pregnancy. Q Rev Biol 68:495– 532.
- Jääskeläinen A, Pussinen J, Nuutinen O, Schwab U, Pirkola J, Kolehmainen M, Jarvelin MR, Laitinen J. 2011. Intergenerational transmission of overweight among Finnish adolescents and their parents: a 16-year follow-up study. Int J Obes (Lond) 35:1289–1294.
- Khulan B, Cooper WN, Skinner BM, Bauer J, Owens S, Prentice AM, Belteki G, Constancia M, Dunger D, Affara NA. 2012. Periconceptional maternal micronutrient supplementation is associated with widespread gender related changes in the epigenome: a study of a unique resource in the Gambia. Hum Mol Genet 21:2086–2101.
- Kivimaki M, Lawlor DA, Smith GD, Elovainio M, Jokela M, Keltikangas-Jarvinen L, Viikari JS, Raitakari OT. 2007. Substantial intergenerational increases in body mass index are not explained by the fetal overnutrition hypothesis: the Cardiovascular Risk in Young Finns Study. Am J Clin Nutr 86:1509–1514.
- Knoll N, Jarick I, Volckmar AL, Klingenspor M, Illig T, Grallert H, Gieger C, Wichmann HE, Peters A, Wiegand S, Biebermann H, Fischer-
Posovszky P, Wabitsch M, Völzke H, Nauck M, Teumer A, Rosskopf D, Rimmbach C, Schreiber S, Jacobs G, Lieb W, Franke A, Hebebrand J, Hinney A. 2014. Mitochondrial DNA variants in obesity. PLoS One 9: e94882.
- Lawlor DA, Smith GD, O'Callaghan M, Alati R, Mamun AA, Williams GM, Najman JM. 2007. Epidemiologic evidence for the fetal overnutrition hypothesis: findings from the mater-university study of pregnancy and its outcomes. Am J Epidemiol 165:418–424.
- Lawlor DA, Timpson NJ, Harbord RM, Leary S, Ness A, McCarthy MI, Frayling TM, Hattersley AT, Smith GD. 2008. Exploring the developmental overnutrition hypothesis using parental–offspring associations and FTO as an instrumental variable. PLoS Med 5:e33.
- Leary S, Davey Smith G, Ness A. 2010. No evidence of large differences in mother-daughter and father-son body mass index concordance in a large UK birth cohort. Int J Obes (Lond) 34:1191–1192.
- Linabery AM, Nahhas RW, Johnson W, Choh AC, Towne B, Odegaard AO, Czerwinski SA, Demerath EW. 2013. Stronger influence of maternal than paternal obesity on infant and early childhood body mass index: the Fels Longitudinal Study. Pediatr Obes 8:159–169.
- Morandi A, Meyre D, Lobbens S, Kleinman K, Kaakinen M, Rifas-Shiman SL, Vatin V, Gaget S, Pouta A, Hartikainen AL, Laitinen J, Ruokonen A, Das S, Khan AA, Elliott P, Maffeis C, Gillman MW, Järvelin MR, Froguel P. 2012. Estimation of newborn risk for child or adolescent obesity: lessons from longitudinal birth cohorts. PloS One 7:e49919.
- Northstone K, Golding J, Davey Smith G, Miller LL, Pembrey M. 2014. Prepubertal start of father's smoking and increased body fat in his sons: further characterisation of paternal transgenerational responses. Eur J Hum Genet 22:1382–1386.
- Parsons TJ, Power C, Logan S, Summerbell CD. 1999. Childhood predictors of adult obesity: a systematic review. Int J Obes Relat Metab Disord 23(Suppl 8):S1–107.
- Patel R, Martin RM, Kramer MS, Oken E, Bogdanovich N, Matush L, Smith GD, Lawlor DA. 2011. Familial associations of adiposity: findings from a cross-sectional study of 12,181 parental-offspring trios from Belarus. PloS One 6:e14607.
- Patro B, Liber A, Zalewski B, Poston L, Szajewska H, Koletzko B. 2013. Maternal and paternal body mass index and offspring obesity: a systematic review. Ann Nutr Metab 63:32–41.
- Pembrey ME, Bygren LO, Kaati G, Edvinsson S, Northstone K, Sjostrom M, Golding J, Team AS. 2006. Sex-specific, male-line transgenerational responses in humans. Eur J Hum Genet 14:159–166.
- Perez-Pastor EM, Metcalf BS, Hosking J, Jeffery AN, Voss LD, Wilkin TJ. 2009. Assortative weight gain in mother-daughter and father-son pairs: an emerging source of childhood obesity. Longitudinal study of trios (EarlyBird 43). Int J Obes (Lond) 33:727–735.
- Pomeroy E, Wells JC, Cole TJ, O'Callaghan M, Stock JT. 2015. Relationships of maternal and paternal anthropometry with neonatal body size, proportions and adiposity in an Australian cohort. Am J Phys Anthropol 156:625–636.
- Power C, Pouliou T, Li L, Cooper R, Hypponen E. 2011. Parental and offspring adiposity associations: insights from the 1958 British birth cohort. Ann Hum Biol 38:390–399.
- Reilly JJ, Armstrong J, Dorosty AR, Emmett PM, Ness A, Rogers I, Steer C, Sherriff A. 2005. Early life risk factors for obesity in childhood: cohort study. BMJ 330:1357.
- Rubin D. 1987. Multiple imputation for nonresponse in surveys. New York: Wiley.
- Schousboe K, Willemsen G, Kyvik KO, Mortensen J, Boomsma DI, Cornes BK, Davis CJ, Fagnani C, Hjelmborg J, Kaprio J, de Lange M, Luciano M, Martin NG, Pedersen N, Pietiläinen KH, Rissanen A, Saarni S, Sørensen TIA, van Baal GCM, Harris JR. 2003. Sex differences in heritability of BMI: a comparative study of results from twin studies in eight countries. Twin Res 6:409–421.
- Speakman JR, Djafarian K, Stewart J, Jackson DM. 2007. Assortative mating for obesity. Am J Clin Nutr 86(2):316–323.
- Steer CD. 2009. Comment on "Clemons T. A look at the inheritance of height using regression toward the mean." Hum Biol 2000(72): 447–454.
- Svensson V, Jacobsson JA, Fredriksson R, Danielsson P, Sobko T, Schioth HB, Marcus C. 2011. Associations between severity of obesity in childhood and adolescence, obesity onset and parental BMI: a longitudinal cohort study. Int J Obes (Lond) 35:46–52.
- Team RDC. 2011. R: A language and environment for statistical computing. Vienna, Austria: R Foundation for Statistical Computing.
- Ulijaszek SJ, Kerr DA. 1999. Anthropometric measurement error and the assessment of nutritional status. Br J Nutr 82:165–177.
- van Buuren S. 2012. Flexible imputation of missing data. Boca Raton, FL: Chapman & Hall/CRC Press.
- van Buuren S, Groothuis-Oudshoorn K. 2011. Multivariate Imputation by Chained Equations in R. J Statistical Softw 45:1–67.
- Veena SR, Krishnaveni GV, Karat SC, Osmond C, Fall CH. 2012. Testing the fetal overnutrition hypothesis; the relationship of maternal and paternal adiposity to adiposity, insulin resistance and cardiovascular risk factors in Indian children. Public health nutrition:1–11.
- Wells JC. 2000. A Hattori chart analysis of body mass index in infants and children. Int J Obes Relat Metab Disord 24:325–329.
- Wells JC. 2014. Commentary: paternal and maternal influences on offspring phenotype: the same, only different. Int J Epidemiol 43:772–774.
- Wells JC, Charoensiriwath S, Treleaven P. 2011. Reproduction, aging, and body shape by three-dimensional photonic scanning in Thai men and women. Am J Hum Biol 23:291–298.
- Wells JC, Chomtho S, Fewtrell MS. 2007. Programming of body composition by early growth and nutrition. Proc Nutr Soc 66(3):423–434.
- Wells JC, Cole TJ. 2014. Height, adiposity and hormonal cardiovascular risk markers in childhood: how to partition the associations? Int J Obes (Lond) 38:930–935.
- Wells JC, Fewtrell MS, Williams JE, Haroun D, Lawson MS, Cole TJ. 2006. Body composition in normal weight, overweight and obese children: matched case-control analyses of total and regional tissue masses, and body composition trends in relation to relative weight. Int J Obes (Lond) 30:1506–1513.
- Wells JC, Williams JE, Chomtho S, Darch T, Grijalva-Eternod C, Kennedy K, Haroun D, Wilson C, Cole TJ, Fewtrell MS. 2012. Bodycomposition reference data for simple and reference techniques and a

4-component model: a new UK reference child. Am J Clin Nutr 96: 1316–1326.

- Whitaker KL, Jarvis MJ, Beeken RJ, Boniface D, Wardle J. 2010. Comparing maternal and paternal intergenerational transmission of obesity risk in a large population-based sample. Am J Clin Nutr 91: 1560–1567.
- Yang Q, Kim SK, Sun F, Cui J, Larson MG, Vasan RS, Levy D, Schwartz F. 2007. Maternal influence on blood pressure suggests involvement of mitochondrial DNA in the pathogenesis of hypertension: the Framingham Heart Study. J Hypertens 25:2067–2073.